

INHOMOGENEOUS RANDOM GRAPHS AND THEIR SCALING LIMITS

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A model of inhomogeneous random graphs

- ▶ For $n \geq 1$, let $\mathbf{w} = (w_1, w_2, \dots, w_n) \in (0, \infty)^n$: $w_i =$ **weight** of vertex i .

For $r \geq 0$, denote $\sigma_r(\mathbf{w}) = \sum_{i=1}^n w_i^r$.

- ▶ Let $\mathcal{G}_{\mathbf{w}}$ be the graph with n vertices (labelled $1, 2, \dots, n$) and where

$(\mathbf{1}_{\text{edge } \{i,j\} \in \mathcal{G}_{\mathbf{w}}})_{1 \leq i < j \leq n}$ are **independent** and have probabilities $f\left(\frac{w_i w_j}{\sigma_1(\mathbf{w})}\right)$,

where $f : \mathbb{R}_+ \rightarrow [0, 1]$ and $f(x) = x + \mathcal{O}(x^2)$, $x \rightarrow 0$.

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e.g. $f(x) = 1 - e^{-x}$: Poisson random graph Aldous '97, Norros & Reittu '05

$f(x) = x \wedge 1$: Chung & Lu '02

$f(x) = x/(1+x)$: Britton & Deijfen & Martin-Löf '06

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Special case (homogenous case): (w_i) all **equal**, then $\mathcal{G}_{\mathbf{w}} = G(n, p)$.

In general, $\deg(i) \approx \text{Poisson}(w_i)$.

Scaling limits of $\mathcal{G}_{\mathbf{w}}$

- ▶ $\mathcal{G}_{\mathbf{w}}^k \equiv k$ -th largest connected components of $\mathcal{G}_{\mathbf{w}}$.
- ▶ Each $\mathcal{G}_{\mathbf{w}}^k$ is a metric space equipped with $d_{\text{gr}} = \text{graph distance of } \mathcal{G}_{\mathbf{w}}$.

Aim: Under suitable **conditions on (\mathbf{w}_n)** , find $\epsilon_n \rightarrow 0$ such that

$$\left(\mathcal{G}_{\mathbf{w}}^k, \epsilon_n \cdot d_{\text{gr}} \right)_{k \geq 1} \xrightarrow[n \rightarrow \infty]{(d)} (\mathbf{G}_k)_{k \geq 1}$$

in certain topology, where \mathbf{G}_k , $k \geq 1$, are some compact (non trivial) metric spaces.

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[Janson '09] **Asymptotic equivalence of random graphs:** The choice of f is irrelevant.

From now on, take $f(x) = 1 - e^{-x}$.

Outline

~> Encoding \mathcal{G}_w with stochastic processes

~> Convergence of the coding processes

~> Identifying the limit graphs.

Outline

- ▶ A LIFO queue representation of \mathcal{G}_w
 \rightsquigarrow Encoding \mathcal{G}_w with stochastic processes
- ▶ Embed \mathcal{G}_w into Galton–Watson trees
 \rightsquigarrow Convergence of the coding processes
- ▶ Construction of the graphs from the limit coding processes
 \rightsquigarrow Identifying the limit graphs.

A LIFO queue representation of \mathcal{G}_w

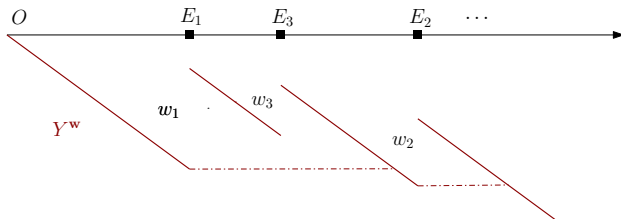
A Last-In-First-Out queueing system:

- ▶ a single server
- ▶ n clients (labelled $1, \dots, n$): Client i arrives at time E_i and requests service time w_i .
- ▶ clients in the queue are served in Last-In-First-Out order: whenever a new customer arrives, the server interrupts the current service (if any) and serves the new-comer. When the latter quits the queue, the server then resumes the previous service.

LIFO queue & coding functions

Define $Y_t^w \equiv -t + \sum_{1 \leq i \leq n} w_i \mathbf{1}_{[0,t]}(E_i)$. Then

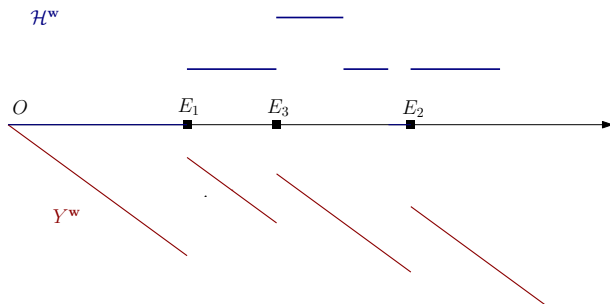
$$Y_t^w - \inf_{s \leq t} Y_s^w = \text{load of the server at } t.$$



LIFO queue & coding functions

$\mathcal{H}_t^w \equiv$ length of the queue at time t . By the LIFO rule, we have

$$\forall t \geq 0: \quad \mathcal{H}_t^w = \#\left\{s \leq t: \inf_{u \in [s, t]} Y_u^w > Y_{s-}^w\right\}.$$



LIFO queue & trees

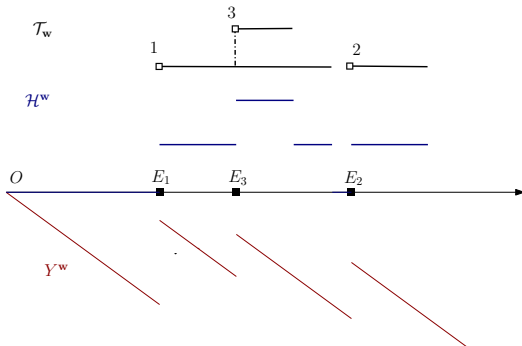
LIFO queue \rightsquigarrow a sequence of family trees \mathcal{T}_w :

- ▶ Vertices of $\mathcal{T}_w = \{ \text{labels of the } n \text{ clients} \}$.
- ▶ Children of $i = \text{labels of the clients interrupting the service of Client } i$.

We have

$\mathcal{H}_t^w = \text{height of } V_t$, where $V_t \equiv \text{label of the client served at time } t$.

In particular, each excursion of \mathcal{H}^w above 0 encodes a tree component in \mathcal{T}_w .



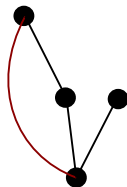
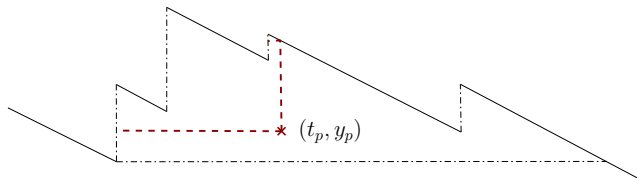
LIFO queue & graph \mathcal{G}_w

To sample some additional edges, let

$$\mathcal{P}^w = \sum_{p \geq 1} \delta_{(t_p, y_p)}$$

be a Poisson point measure on \mathbb{R}_+^2 of intensity $\frac{1}{\sigma_1(w)} \mathbf{1}_{\{0 \leq x \leq Y_t^w - \inf_{[0,t]} Y^w\}} dt dx$. Set

$$\mathcal{A}^w = \{ \{V_{s_p}, V_{t_p}\} : 1 \leq p \leq |\mathcal{P}^w| \} \quad \text{where} \quad s_p \equiv \inf \left\{ s : \inf_{s \leq u \leq t_p} Y_u^w - \inf_{s \leq t} Y_s^w > y_p \right\}$$



LIFO queue & graph $\mathcal{G}_{\mathbf{w}}$

Denote by $\tilde{\mathcal{G}} = \mathcal{T}_{\mathbf{w}} \cup \mathcal{A}^{\mathbf{w}}$.

Theorem. Suppose $\frac{w_i}{\sigma_1(\mathbf{w})} E_i, 1 \leq i \leq n$, are i.i.d. $\mathbf{Exp}(1)$. Then $\tilde{\mathcal{G}} \stackrel{(d)}{=} \mathcal{G}_{\mathbf{w}}$.

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In consequence, $\mathcal{G}_{\mathbf{w}}$ is encoded by $Y^{\mathbf{w}}$, $\mathcal{H}^{\mathbf{w}}$ and $\mathcal{P}^{\mathbf{w}}$. Therefore, to find the scaling limit of $\mathcal{G}_{\mathbf{w}}$, we look at

- ▶ the convergence of $Y^{\mathbf{w}}$: ✓
- ▶ the convergence of $\mathcal{P}^{\mathbf{w}}$: ✓
- ▶ the convergence of $\mathcal{H}^{\mathbf{w}}$: not a continuous funct. of $Y^{\mathbf{w}}$...

Preview of the embedding

To embed \mathcal{G}_w into Galton–Watson trees,

- ▶ introduce a Markovian LIFO queue \rightsquigarrow Galton–Watson trees
- ▶ embed the previous n -client queue into the Markovian one.

Markovian LIFO queue & GW trees

From now on, **suppose** $\sigma_2(\mathbf{w}) \leq \sigma_1(\mathbf{w})$.

A Markovian LIFO queueing system:

- ▶ A single server treats the clients in Last-In-First-Out order
- ▶ An ∞ -sequence of clients arrive at rate 1:
 - ▶ the k -th client arrives at time τ_k ,
 - ▶ chooses his type $J_k \in \{1, \dots, n\}$ with probabilities $\mathbf{P}(J_k = i) = \frac{w_i}{\sigma_1(\mathbf{w})}$
 - ▶ requests service time w_i if $J_k = i$.

■ Load of the server: $X_t^{\mathbf{w}} \equiv -t + \sum_{k \geq 1} w_{J_k} \mathbf{1}_{[0, t]}(\tau_k)$.

$X^{\mathbf{w}}$ is a Lévy process of Lévy measure $\nu_{\mathbf{w}} \equiv \frac{1}{\sigma_1(\mathbf{w})} \sum_i w_i \delta_{w_i}$ and ≤ 0 drift.

■ As before, we can associate with the queue a sequence of family trees $\mathbf{T}_{\mathbf{w}}$.

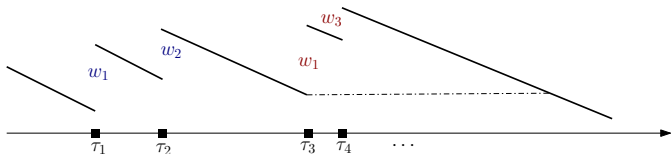
It turns out $\mathbf{T}_{\mathbf{w}} = \text{i.i.d. Galton-Watson trees with offspring distribution } \mathbf{Poisson}(W_n), \text{ where } W_n \sim \nu_{\mathbf{w}}.$

■ $\mathbf{H}^{\mathbf{w}} \equiv \text{length of the queue} = \text{height process of } \mathbf{T}_{\mathbf{w}}.$

Embedding of the n -client queue

Color the clients in **blue** or **red** according to the following rule:

- ▶ If the type of the k -th client $\in \{\text{types of previous blue clients}\}$, the client is in **red**;
- ▶ otherwise, the client inherits his color from his parent, with the convention that if there is no parent, the client is in **blue**.



Embedding of the n -client queue

$\mathcal{B} \equiv \{t \geq 0 : \text{either a blue client is served at } t \text{ or the server is idle at } t\}$

Let

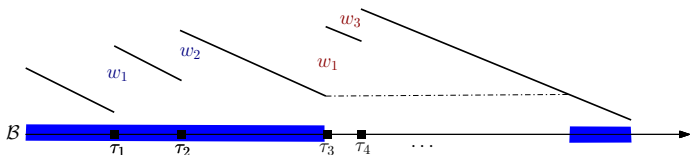
$X^w \circ \theta^w =$ process obtained by **restricting** X^w to \mathcal{B} ,

namely, $\Lambda_t^w \equiv \int_0^t \mathbf{1}_{\mathcal{B}}(s) ds$ and then $\theta_t^w \equiv \inf\{s : \Lambda_s^w > t\}$.

Lemma. We have

$$X^w \circ \theta^w \stackrel{(d)}{=} Y^w, \quad H^w \circ \theta^w \stackrel{(d)}{=} \mathcal{H}^w.$$

Problem: difficult to prove directly the convergence of θ^w .



Another description of θ^w

Recall the k th client arrives at τ_k and is of type J_k . Let

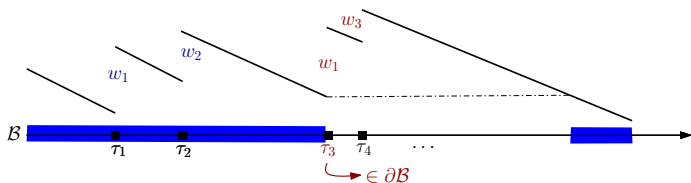
$$X_t^{b,w} \equiv X_{\theta_t^w}^w + \overbrace{\sum_{k \geq 1} w_{J_k} \mathbf{1}_{\{\tau_k \in \partial B\}} \mathbf{1}_{\{\wedge_{\tau_k}^{b,w} \leq t\}}}^{A_t^w} \quad \text{and} \quad X^{r,w} \equiv X^w - X^{b,w}.$$

Lemma. $X^{b,w}, X^{r,w}$ are two independent copies of X^w . Moreover,

$$\theta_t^w = t + \gamma_{A_t^w}^w,$$

where $\gamma_x^w = \inf\{t : X_s^{r,w} < -x\}$ and

$$A_t^w = \sum_i w_i (N_i(t) - 1)_+, \quad \text{where} \quad N_i(t) = \#\{\text{jumps of } X^{b,w} \text{ of type } i \text{ before } t\}$$



Construction of $\mathcal{G}_{\mathbf{w}}$ from Lévy processes

- ▶ **Sample $X^{b,\mathbf{w}}, X^{r,\mathbf{w}}$.** Let $X^{b,\mathbf{w}}, X^{r,\mathbf{w}}$ be two independent copies of a (sub-)critical spectrally positive Lévy process such that

$$\psi_n(\lambda) \equiv \log \mathbb{E}[e^{-\lambda X_1^b}] = \left(1 - \frac{\sigma_2(\mathbf{w})}{\sigma_1(\mathbf{w})}\right)\lambda + \int (e^{-\lambda u} - 1 + \lambda u)\nu_{\mathbf{w}}(du),$$

where $\nu_{\mathbf{w}} = \frac{1}{\sigma_1(\mathbf{w})} \sum_{i \geq 1} w_i \delta_{w_i}$. Set $X^{\mathbf{w}} \equiv X^{b,\mathbf{w}} + X^{r,\mathbf{w}}$.

- ▶ **Define $Y^{\mathbf{w}}$.** Let

$$Y^{\mathbf{w}} \equiv X^{b,\mathbf{w}} - A^{\mathbf{w}}, \quad \text{where} \quad A_t^{\mathbf{w}} \equiv \sum_i w_i (N_i(t) - 1)_+.$$

- ▶ **Define $\theta^{\mathbf{w}}$.** Set $\theta_t^{\mathbf{w}} = t + \gamma_{A_t^{\mathbf{w}}}^{\mathbf{w}}$, where $\gamma_x^{\mathbf{w}} = \inf\{t : X_t^{r,\mathbf{w}} < -x\}$.

- ▶ **Define $\mathcal{H}^{\mathbf{w}}$.** Let

$$\mathcal{H}^{\mathbf{w}} = \mathbf{H}^{\mathbf{w}} \circ \theta^{\mathbf{w}} \quad \text{where} \quad \mathbf{H}_t^{\mathbf{w}} = \#\{s \leq t : X_{s-}^{\mathbf{w}} < \inf_{[s,t]} X^{\mathbf{w}}\}.$$

- ▶ **Sample additional edges** according to $\mathcal{P}^{\mathbf{w}}$, which is a Poisson point measure on \mathbb{R}_+^2 of intensity $\frac{1}{\sigma_1(\mathbf{w})} \mathbf{1}_{\{0 \leq x \leq Y_t^{\mathbf{w}} - \inf_{[0,t]} Y^{\mathbf{w}}\}} dt dx$.

Construction of the limit graph: Part I

- ▶ **Sample X^b, X^r .** Let X^b, X^r be two independent copies of a (sub-)critical spectrally positive Lévy process such that

$$\psi(\lambda) \equiv \log \mathbb{E}[e^{-\lambda X_1^b}] = \alpha\lambda + \frac{1}{2}\beta\lambda^2 + \int \kappa(e^{-\lambda u} - 1 + \lambda u)\pi(du),$$

where $\alpha, \beta \geq 0$, $\kappa > 0$ and $\pi = \sum_{j \geq 1} c_j \delta_{c_j}$ with $c_1 \geq c_2 \geq \dots \geq 0$ satisfying $\sum_i c_i^3 < \infty$. Namely,

$$X_t^b = -\alpha t + \sqrt{\beta} B_t + \sum_{i \geq 1} c_i (N_i(t) - c_i \kappa t),$$

where B = standard Brownian Motion and N_i = Poisson process of rate κc_i .

- ▶ **Define Y .** Let

$$Y \equiv X^b - A, \quad \text{where} \quad A_t \equiv \frac{1}{2}\beta\kappa t^2 + c_i(N_i(t) - 1)_+.$$

- ▶ **Define θ .** Set $\theta_t = t + \gamma_{A_t}$, where $\gamma_x = \inf\{t : X_t^r < -x\}$.

Construction of the limit graph: Part II

- ▶ **Height process of a Lévy process.** For the Lévy process X , we **can** define an analogue of the discrete height process \mathbf{H}^w .

Le Gall & Le Jan '98: Suppose $\int^\infty d\lambda/\psi(\lambda) < \infty$. Then there exists a continuous process \mathbf{H} such that

$$\mathbf{H}_t = \lim_{\epsilon \rightarrow 0} \frac{1}{\epsilon} \int_0^t \mathbf{1}_{\{X_s - \inf_{[s,t]} X < \epsilon\}} \quad \text{exists in probability}$$

\mathbf{H} is in fact the **height process** of the Lévy tree with branching mechanism ψ .

- ▶ **Define \mathcal{H} .** Let

$$\mathcal{H} \equiv \mathbf{H} \circ \theta,$$

- ▶ **Sample additional edges** according to \mathcal{P} , which is a Poisson point measure on \mathbb{R}_+^2 of intensity $\mathbf{1}_{\{x \leq Y_t - \inf_{[0,t]} Y\}} dt dx$.

Convergence of the graphs $\mathcal{G}_{\mathbf{w}_n}$

Theorem. Let $\mathbf{w}_n = (w_{n1}, \dots, w_{nn})$, $n \geq 1$. Denote $\psi_n(\lambda) = \log \mathbf{E}[e^{-\lambda X_1^{\mathbf{w}_n}}]$. Suppose

$$a_n \rightarrow \infty, \quad b_n/a_n \rightarrow \infty \quad \text{and} \quad b_n/a_n^2 \rightarrow \beta_0 \in [0, \beta] \quad \text{satisfying that}$$

$$(1) \quad \forall \lambda \geq 0: \quad b_n \psi_n(\lambda/a_n) \rightarrow \psi(\lambda) \quad \text{and} \quad \lim_{y \rightarrow \infty} \limsup_{n \rightarrow \infty} \frac{a_n}{b_n} \int_{y/a_n}^1 \frac{d\lambda}{\psi_n(\lambda)} = 0.$$

Then we have

$$(2) \quad \left(\left(\frac{1}{a_n} \mathcal{Y}_{b_n t}^{\mathbf{w}_n} \right)_{t \geq 0}, \left(\frac{a_n}{b_n} \mathcal{H}_{b_n t}^{\mathbf{w}_n} \right)_{t \geq 0}, \tilde{\mathcal{P}}^{\mathbf{w}_n} \right) \xrightarrow{d} (Y, \mathcal{H}, \mathcal{P}) \quad \text{in} \quad \mathbb{D} \times \mathbb{C} \times \mathcal{M}(\mathbb{R}_+^2),$$

where $\langle f, \tilde{\mathcal{P}}^{\mathbf{w}_n} \rangle = \sum_{\rho \geq 1} f\left(\frac{t\rho}{b_n}, \frac{y\rho}{a_n}\right)$, $\forall f: \mathbb{R}_+^2 \rightarrow \mathbb{R}_+$ measurable.

Essential ingredient of the proof:

[Duquesne & Le Gall '02](#) Under Condition (1), we have the convergence of the Poisson(W_n)-Galton–Watson trees to ψ -Lévy trees.

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We can deduce from (2) the convergence of the graph $\mathcal{G}_{\mathbf{w}_n}$:

$$(\mathcal{G}_{\mathbf{w}_n}^k, \frac{a_n}{b_n} d_{\text{gr}})_{k \geq 1} \xrightarrow{d} (\mathbf{G}_k)_{k \geq 1} \quad \text{in Gromov–Hausdorff topology,}$$

where \mathbf{G}_k , $k \geq 1$, are the connected components of the limit graph constructed from $(Y, \mathcal{H}, \mathcal{P})$.

Convergence of the graphs \mathcal{G}_{w_n}

From the previous theorem, we can recover

- ▶ Addario-Berry & Broutin & Goldschmidt '12 $G(n, p)$:
(w_{ni}) all equal, then $\psi(\lambda) = \alpha\lambda + \lambda^2/2$.
- ▶ Bhamidi & van der Hofstad & Sen '17+ Power-law case:
 $w_{ni} \sim (n/i)^\gamma$, $\gamma \in (\frac{1}{3}, \frac{1}{2})$, then $\beta = 0$ and $c_i = i^{-\gamma}$, $i \geq 1$.
Moreover, they show $\dim_h(\mathbf{G}_k) \leq \frac{1-\gamma}{1-2\gamma}$, a.s.

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[Duquesne & Le Gall '05] fractal properties of Lévy trees + our construction of \mathbf{G}_k

Proposition. Let

$$\bar{\gamma} := \inf\{s \geq 0 : \lim_{j \rightarrow \infty} j^s c_j = \infty\} \quad \text{and} \quad \underline{\gamma} \equiv \sup\{s \geq 0 : \lim_{j \rightarrow \infty} j^s c_j = 0\}$$

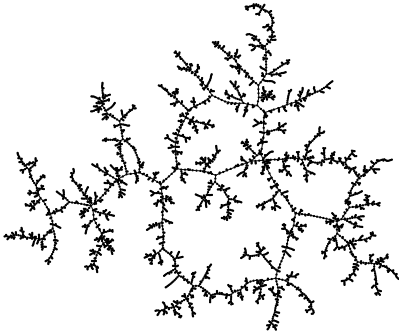
Suppose $\beta = 0$ and $\bar{\gamma} < 1/2$. Then **P**-a.s. for all $k \geq 1$,

$$\dim_p(\mathbf{G}_k) = \frac{1-\underline{\gamma}}{1-2\underline{\gamma}} \quad \text{and} \quad \dim_h(\mathbf{G}_k) = \frac{1-\bar{\gamma}}{1-2\bar{\gamma}}.$$

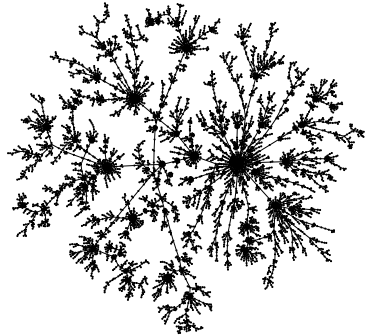
In particular, if (c_j) varies regularly with index γ , then $\dim_p(\mathbf{G}_k) = \dim_h(\mathbf{G}_k) = \frac{1-\gamma}{1-2\gamma}$.

Simulation of large \mathcal{G}_w^k

Homogeneous case



Power law case



THANK YOU!